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Title: Precision sensing assisted by quantum-classical computation

Author(s): Sone, Akira

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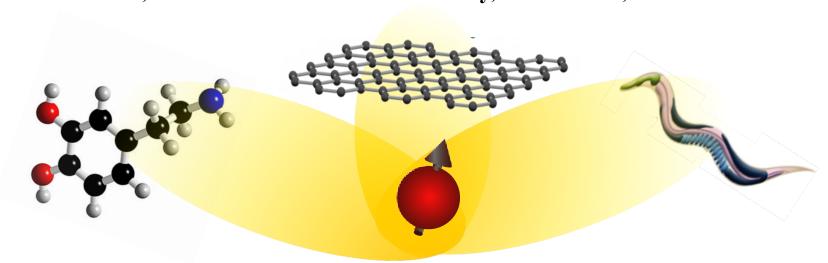
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# Precision sensing assisted by quantum-classical computation

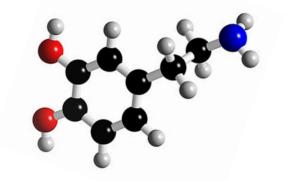
#### **Akira Sone**

Theoretical Division, Los Alamos National Laboratory, Los Alamos, NM 87545 CNLS, Los Alamos National Laboratory, Los Alamos, NM 87545

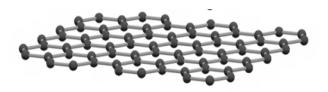


### Quantum sensing for...

Room-temperature NMR



Magnetic material



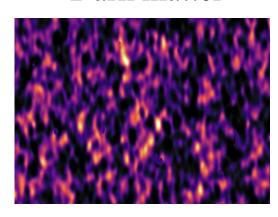
Living system (e.g. C. elegance)



Gravitational wave detection



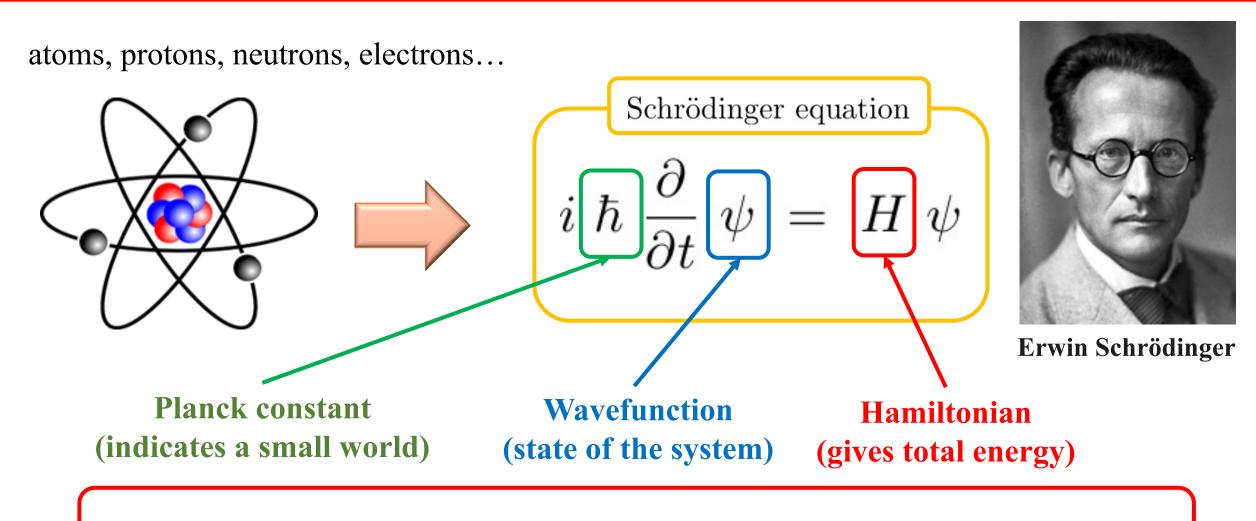
Dark matter



Chemistry, material science, medical science, cosmology...

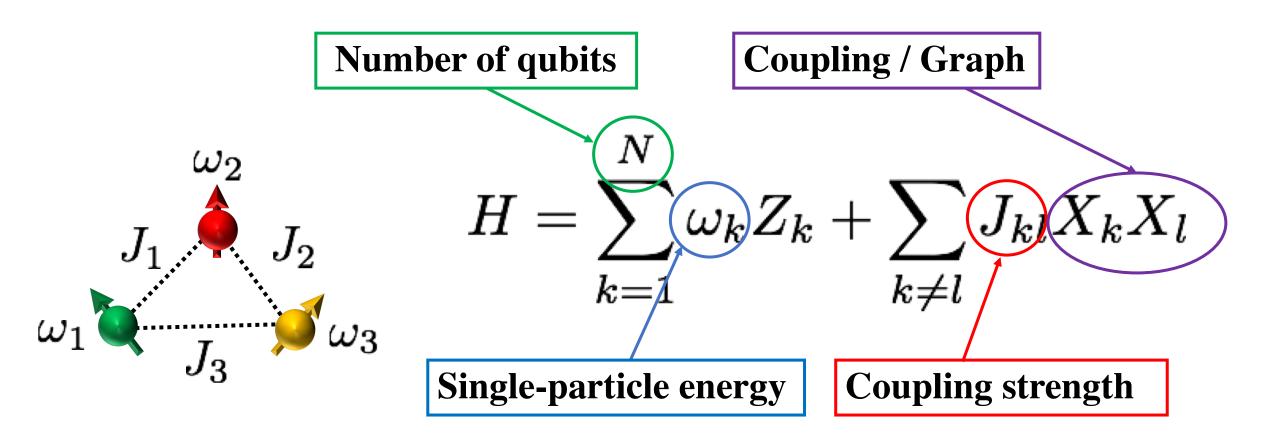
# Practical scenario (not precise, but still interesting...)

#### Example of practical scenario: Molecular structure determination



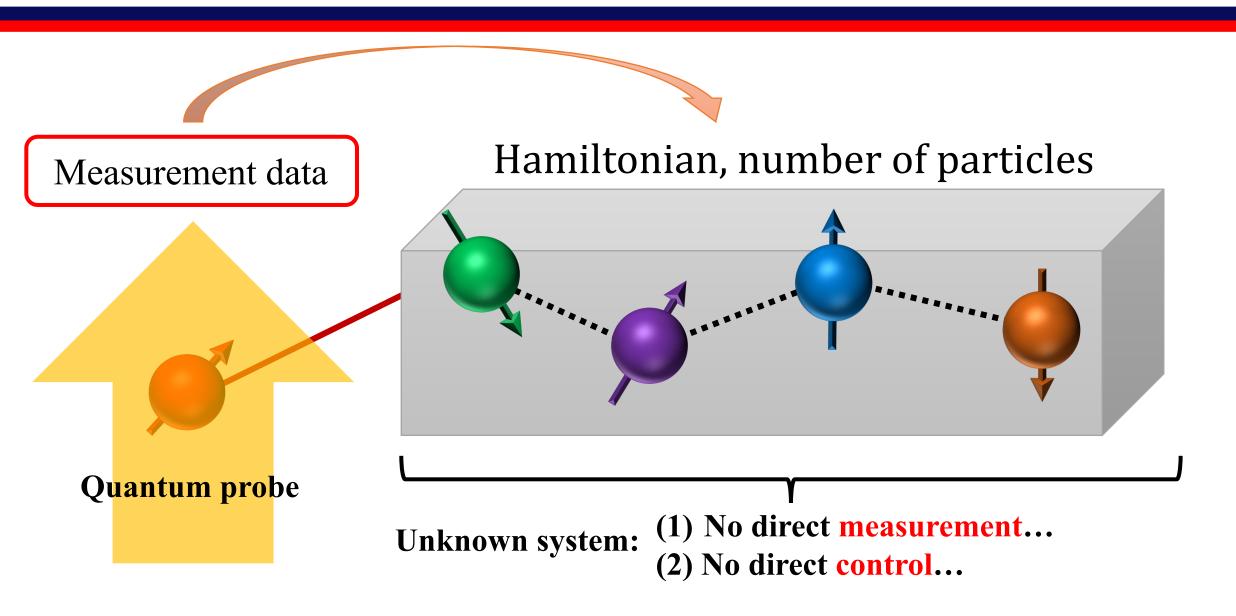
The information of molecule is included in the Hamiltonian

#### Example of practical scenario: Molecular structure determination



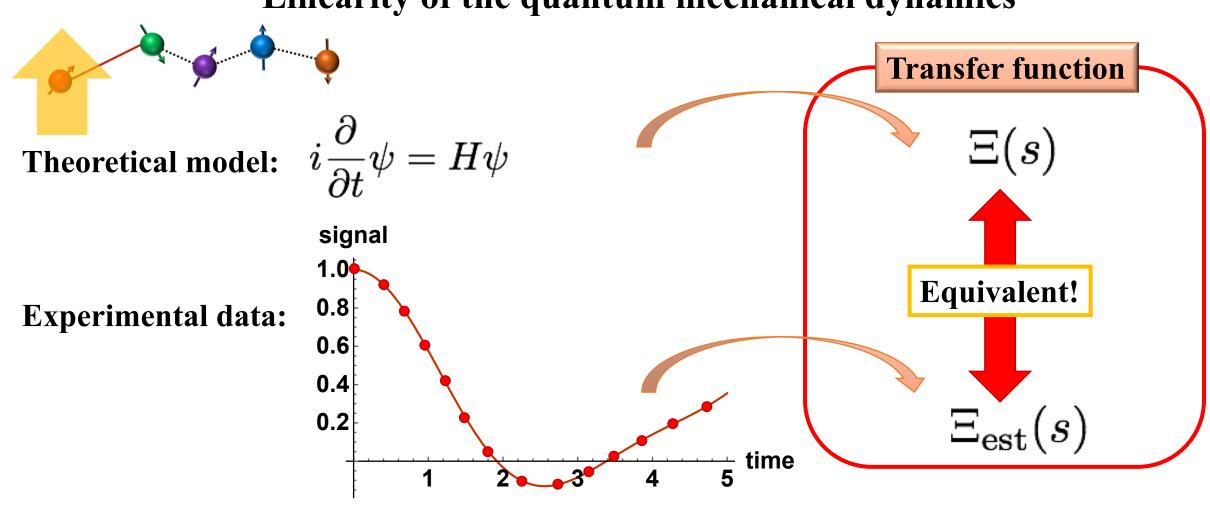
We want to extract this information from measurements

#### Example of practical scenario: Molecular structure determination



#### **Estimation of Hamiltonian parameters:**

#### Linearity of the quantum mechanical dynamics



### **Estimation of Hamiltonian parameters:**

signal

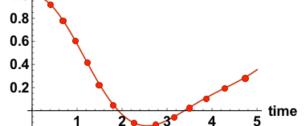
Theoretical model:  $i\frac{\partial}{\partial t}\psi = H\psi$ 

polynomials

$$\Xi(s) = \frac{s^m + g_2(\theta_1, \cdots, \theta_p)s^{m-1} + \cdots}{s^n + g_1(\theta_1, \cdots, \theta_p)s^{n-1} + \cdots}$$

**Experimental data:** 





numbers

$$\Xi_{\text{est}}(s) = \frac{s^m + (a_2)^{m-1} + \cdots}{s^n + (a_1)^{m-1} + \cdots}$$

equal

#### **Estimation of Hamiltonian parameters:**

#### We must have:



$$\Xi(s) = \Xi_{\text{est}}(s)$$

Coefficients of the Laplace variable s in  $\Xi(s)$  are polynomials. Coefficients of the Laplace variable s in  $\Xi_{\rm est}(s)$  are numbers.

#### System of polynomial equations

$$\begin{cases} f_1(\theta_1, \cdots, \theta_p) = 0 \\ f_2(\theta_1, \cdots, \theta_p) = 0 \\ f_3(\theta_1, \cdots, \theta_p) = 0 \\ \vdots \\ f_q(\theta_1, \cdots, \theta_p) = 0 \end{cases}$$



Solve the equation directly

Find the parameters

$$\{\theta_1, \theta_2, \cdots, \theta_p\}$$

# **Precision sensing**

## Quantify ultimate precision

#### **Quantum Fisher information (QFI):**

$$I(\theta; \rho_{\theta}) = 8 \lim_{\epsilon \to 0} \frac{1 - \mathbb{F}[\rho_{\theta}, \rho_{\theta + \epsilon}]}{\epsilon^2}$$

Quantum Cramer-Rao bound:

Fidelity: distance measure of two quantum states

$$\delta heta_{\{\Pi_j\}}^2 \geq \frac{1}{I(\theta; \rho_{\theta})}$$
 Ultimate precision limit

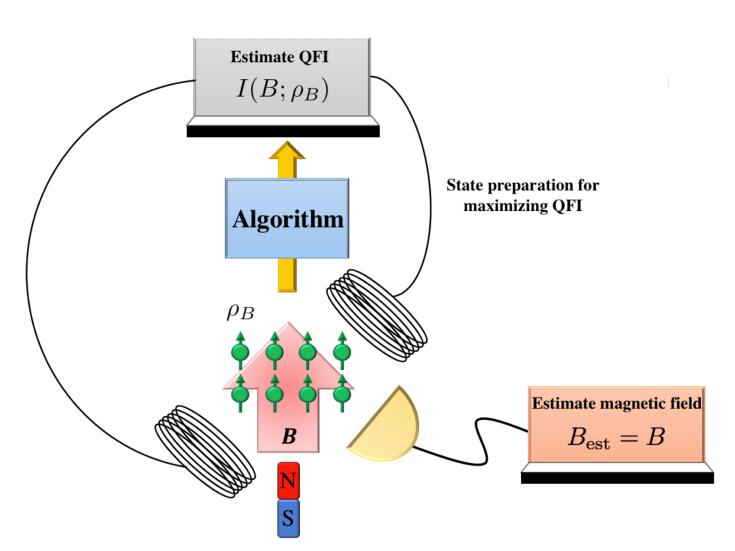
 $\{\Pi_i\}$ : set of measurements

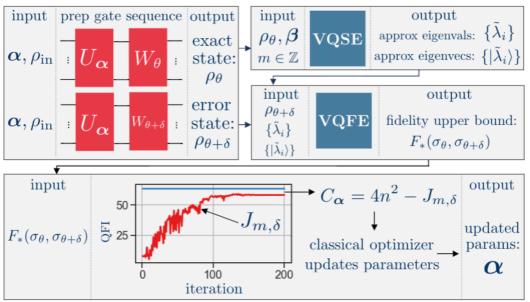
Ultimate precision limit (how good the quantum sensor is)

For the optimal measurement  $\{\Pi_i^*\}$ , the variance satisfies the bound

$$\delta\theta_{\{\Pi_j^*\}}^2 = \delta\theta_{\min} = \frac{1}{I(\theta; \rho_{\theta})}$$

#### Quantum sensing assisted by a quantum-classical computer





- 1. Computing the lower bound of QFI by using the truncated state constructed with smaller number of eigenvalues
- 2. Feedback control to prepare the good state
- 3. Finding optimal measurement

#### Quantum sensing assisted by a quantum-classical computer

As increase the size of truncated state, the lower bound is getting closer to the real QFI

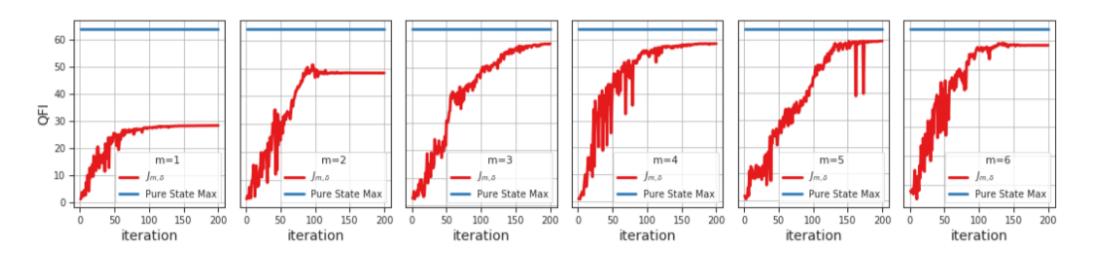


FIG. 3. Figure showing lower bound,  $J_{m,\delta}$ , versus iteration for different m values (number of non-zero eigenvalues kept). A 4-qubit state of purity  $\sim 0.95$  was used to generate this data. The key point is that our lower bound increases with m.

Ongoing research: Finding optimal measurement